Gradually Truncated Log-normal distribution - Size distribution of firms

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Abstract

Many natural and economical phenomena are described through power law or log-normal distributions. In these cases, probability decreases very slowly with step size compared to normal distribution. Thus it is essential to cut-off these distributions for larger step size. Recently we introduce the gradually truncated power law distribution to successfully describe variation of financial, educational, physical and citation index. In the present work, we introduce gradually truncated log-normal distribution in which we gradually cut-off larger steps due to physical limitation of the system. We applied this distribution successfully to size distribution of USA's manufactoring firms which is measured through their annual sell. The physical limitation are due to limited market size or shortage of highly competent executives.

I. Introduction

Many natural [1-16] and economical [17-23] phenomena are distributed through power law [24,25] or log-normal [26] distribution. In log-normal distribution, the logarithms of the larger steps are exponentially rare in contrast to a normal distribution in which the larger steps themselves are exponentially rare [27]. Thus the probability distribution decreases very slowly with step size in compare to normal distribution as in power law distribution. Recently we introduced gradually truncated power law [23,28,29] in which we

combine a statistical distribution factor and a gradual cut-off after a certain step size which is due to the limited physical capacity of the system to analyze the complex systems. In the present paper we consider that larger steps of log-normal distribution should also be gradually cut-off after certain step size like in power law distribution. Thus, we propose a gradually truncated log-normal distribution, which in line with gradually truncated power law [23] or Lévy distribution [28] is described through:

$$P(\ln x) = \frac{1}{\sqrt{2\pi}\sigma} \exp(-\frac{(\ln x - \mu)^2}{2\sigma^2}) f(x) \tag{1}$$

where μ and σ are the mean and standard deviation of $\ln x$. Further

$$f(x) = \begin{cases} 1 & \text{if } |x| \leqslant x_c \\ -\left(\frac{|x| - x_c}{k}\right)^{\beta} & \text{if } |x| > x_c \end{cases}$$
 (2)

where x_C is some cut-off value where physical limitations of the system become important. k and β are constants of the system for this factor. Though Central Limit theorem all distribution must approach to normal distribution [27]. For log-normal distribution to approach normal distribution for larger steps, we should have:

$$\beta = 2 \tag{3}$$

Thus:

$$f(x) = \begin{cases} 1 & \text{if } |x| \leqslant x_c \\ -\left(\frac{|x|-x_c}{k}\right)^2 \end{cases} & \text{if } |x| > x_c \end{cases} \tag{4}$$

Generally, there is too little mass in the upper tail of the distribution. However the statistical technique known as Zipf plot [30,31] which is a plot of the log of the rank versus the log of the variable, is very useful to discuss the distribution in the upper tail correctly

Let $(x_1, x_2, ..., x_N)$ be a set of N observations on a random variable x and supose that the observations are ordered from largest to smallest so that the index i is in the rank of x_i . The Zipf plot of the sample is the graph of $\ln x_i$ against $\ln i$.

Thus x_i may be estimated by the criterion [32]

$$\int_{x_i}^{\infty} N(x)dx = \int_{x_i}^{\infty} N.P(x)dx = i$$
 (5)

This specifies that there are i number out of the ensemble N which are equal or more than x_i . From the dependence of x_i on i in a Zipf plot, one can test whether it accord with a hypothesised form for P(x). It accentuates the upper tail of the distribution and therefore make it easier to detect deviations in the upper tail from the theoretical prediction of a particular distribution. Using Equation (1) and (5) and considering $P(\ln x)d(\ln x) = P(x)dx$ we get:

$$i = \frac{N}{\sqrt{2\pi}\sigma} \int_{x_i}^{\infty} \frac{e^{-\frac{(\ln x - \mu)^2}{2\sigma^2}}}{x} f(x) dx \tag{6}$$

or

$$\ln i = \ln\left(\frac{N}{\sqrt{2\pi}\sigma} \int_{x_i}^{\infty} \frac{e^{-\frac{(\ln x - \mu)^2}{2\sigma^2}}}{x} f(x) dx\right) + C \tag{7}$$

where C is a constant and is approximately equal to

$$C = \ln(\frac{N}{\sqrt{2\pi}\sigma})\tag{8}$$

Now we discuss this distribution for size distribution of firms. This is important because of their implications for the literature on the dynamics of firm growth. Gibrat [26] showed that if the distribution of growth rates is independent of firm size, the static distribution of firm size would approach the log-normal. Hart and Prais [33] and Hall [34] found evidence using British empirical data that the log-normal fits the distribution of firm size reasonably well with some skewness to the rigth and that the growth rates of firms were independence of initial size. Stanley et. al. [35] using Zipf plot technique showed that although exist a agreement with log distribution there is significant deviation for first 100 firms from this distribution.

Here we take our empirical data from Stanley et. al. [35], which are annual sell of 4071 Compustat firms in SIC codes 2000-3999 in year 1993. From log-normal distribution plot of the sell of these firms, we consider $\mu = 17.76$

and $\sigma=2.72$ as also have been considered by them. We further consider $x_C=8.10^9$ as deviation start from this point and $k=1.5.10^{11}$. In Figure (1) we plot log rank versus log of sales. The empirical result are in good agreement with present theory. For comparison we also plot log-normal distribution. We especulate that this limiting factor come from market capacity. If we consider log-normal distribution, the sale of these firms must be of the order of 30.10^{12} dollars, about three times the Gross national product of USA for this year, which is not possible. Only first ten industries would have sell of the order of 20.10^{12} dollars i.e. about double of Gross National product. The total sales of all industries through our model (gradually truncated log-normal) comes out to be of the order of 10.10^{12} dollars, almost of the order of Gross national product, which seems reasonable considering that: (i) Gross National product includes all services and manufactured good. (ii) All industries do not make final product. Many times final product of one industry is initial product of other.

In conclusion sale of the very large firms depend not only on their inherent potential to grow but also on available market, which however is not the case for small industries, because enough market is available if they can compete. We feel that like in power law distribution, the log-normal distribution should also be gradually truncated due to physical limit of the system.

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Figure Captions

Figure 1 Zipf plot double logarithmic plot of sales versus rank for 4071 Compustat firms in SIC Codes 2000-3999 for year 1993 in USA. represent empirical ,- - - - represent log-normal while — represents gradually truncated log-normal distribution

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